

1 Hand-written Part

P₁

$$\begin{aligned}\varphi'(s) &= \theta(s) + s \theta'(s) \\ &= \theta(s) + s \cdot \frac{e^{-s}}{(1+e^{-s})^2} \\ &= \theta(s) + s \cdot \theta(s) \cdot \frac{e^{-s}}{(1+e^{-s})} \\ &= \theta(s) + s \cdot \theta(s) \cdot (1 - \theta(s)) \\ &= -s \theta^2(s) + (s+1) \theta(s)\end{aligned}$$

P₂

(A)

$$\begin{aligned}v_0 &= \left[\frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right]^T \\ v_1 &= P v_0 = \left[\frac{1}{2}, \frac{1}{6}, \frac{1}{3}\right]^T \\ v_2 &= P v_1 = \left[\frac{1}{3}, \frac{1}{6}, \frac{1}{2}\right]^T \\ v_3 &= P v_2 = \left[\frac{5}{12}, \frac{1}{4}, \frac{1}{3}\right]^T \\ v_4 &= P v_3 = \left[\frac{5}{12}, \frac{1}{6}, \frac{5}{12}\right]^T \\ v_5 &= P v_4 = \left[\frac{3}{8}, \frac{5}{24}, \frac{5}{12}\right]^T\end{aligned}$$

(B)

Let $v^* = [x, y, z]^T$

$$v^* = P v^* \Rightarrow \begin{cases} x = y + \frac{1}{2}z \\ y = \frac{1}{2}z \\ z = x \end{cases}$$

$$2x = y = z$$

$$v^* = \left[\frac{2}{5}, \frac{1}{5}, \frac{2}{5}\right]^T$$

P₃

(A)

$$\begin{aligned}\text{dis}((1,2), (7,0)) &= \sqrt{40} & \begin{array}{c|c} (1,2) & (3,4) \\ \hline (1,2) & (3,4) \\ & (7,0) \\ & (10,2) \end{array} & \begin{array}{c|c} (1,2) & (\frac{20}{3}, 2) \\ \hline (1,2) & (7,0) \\ (3,4) & (10,2) \end{array} & \begin{array}{c|c} (2,3) & (\frac{17}{2}, 1) \\ \hline (1,2) & (7,0) \\ (3,4) & (10,2) \end{array} \\ \text{dis}((3,4), (7,0)) &= \sqrt{32} & & & & \\ \text{dis}((1,2), (10,2)) &= \sqrt{81} & & & & \\ \text{dis}((3,4), (10,2)) &= \sqrt{53} & \begin{array}{c|c} (1,2) & (\frac{20}{3}, 2) \\ \hline (2,3) & (\frac{17}{2}, 1) \end{array} & \begin{array}{c|c} (2,3) & (\frac{17}{2}, 1) \\ \hline (2,3) & (\frac{17}{2}, 1) \end{array} & \text{converge}\end{aligned}$$

(B)

$$\begin{array}{c|c} (1,2) & (7,0) \\ \hline (1,2) & (7,0) \\ (3,4) & (10,2) \\ \hline (2,3) & (\frac{17}{2}, 1) \end{array} \quad \begin{array}{c|c} (2,3) & (\frac{17}{2}, 1) \\ \hline (1,2) & (7,0) \\ (3,4) & (10,2) \\ \hline (2,3) & (\frac{17}{2}, 1) \end{array} \quad \text{converge}$$

No, the result is same as (A)

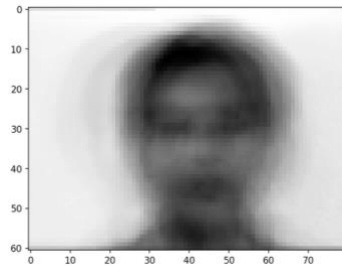
(C)

$$\begin{array}{c|c} (1,2) & (3,4) \\ \hline (1,2) & (3,4) \\ & (5,6) \\ & (7,0) \\ & (10,2) \\ \hline (1,2) & (\frac{25}{4}, 3) \end{array} \quad \begin{array}{c|c} (1,2) & (\frac{25}{4}, 3) \\ \hline (1,2) & (5,6) \\ (3,4) & (7,0) \\ & (10,2) \\ \hline (2,3) & (\frac{22}{3}, \frac{8}{3}) \end{array} \quad \begin{array}{c|c} (2,3) & (\frac{22}{3}, \frac{8}{3}) \\ \hline (1,2) & (5,6) \\ (3,4) & (7,0) \\ & (10,2) \\ \hline (2,3) & (\frac{22}{3}, \frac{8}{3}) \end{array} \quad \begin{array}{c|c} (1,2) & (10,2) \\ \hline (1,2) & (7,0) \\ (3,4) & (10,2) \\ & (5,6) \\ \hline (3,4) & (\frac{17}{2}, 1) \end{array} \quad \begin{array}{c|c} (3,4) & (\frac{17}{2}, 1) \\ \hline (1,2) & (7,0) \\ (3,4) & (10,2) \\ & (5,6) \\ \hline (3,4) & (\frac{17}{2}, 1) \end{array} \quad \begin{array}{c|c} (1,2) & (3,4) \\ \hline & \text{converge to} \\ & (2,3) (\frac{22}{3}, \frac{8}{3}) \\ & \text{which is different from} \\ & (1,2) (10,2) \\ & \text{converge to} \\ & (3,4) (\frac{17}{2}, 1) \end{array}$$

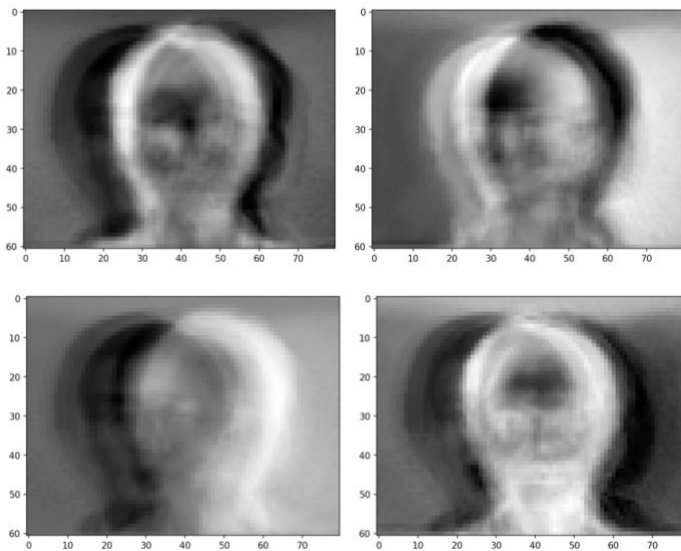
2 Programming Part

(a)

Mean

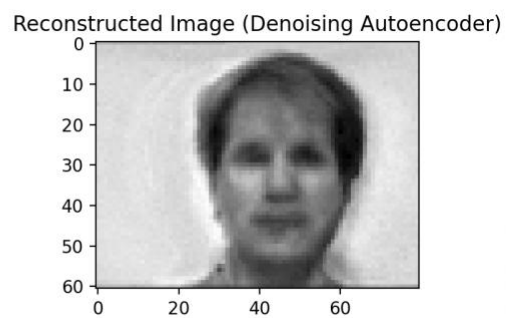
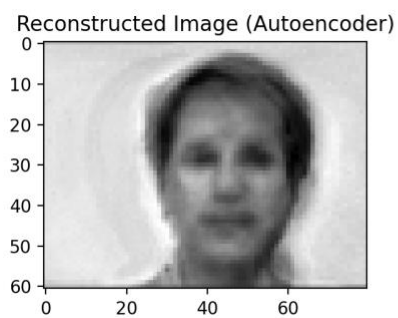
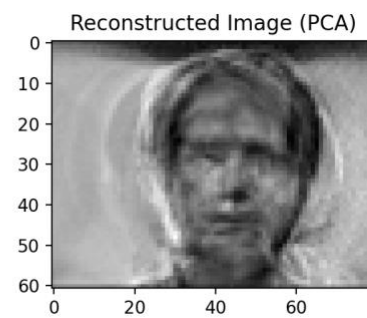
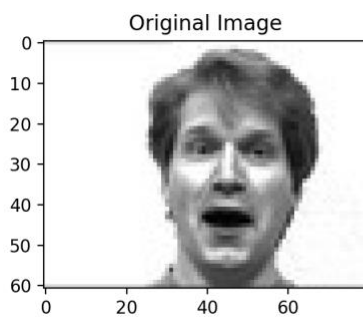


Top 4:

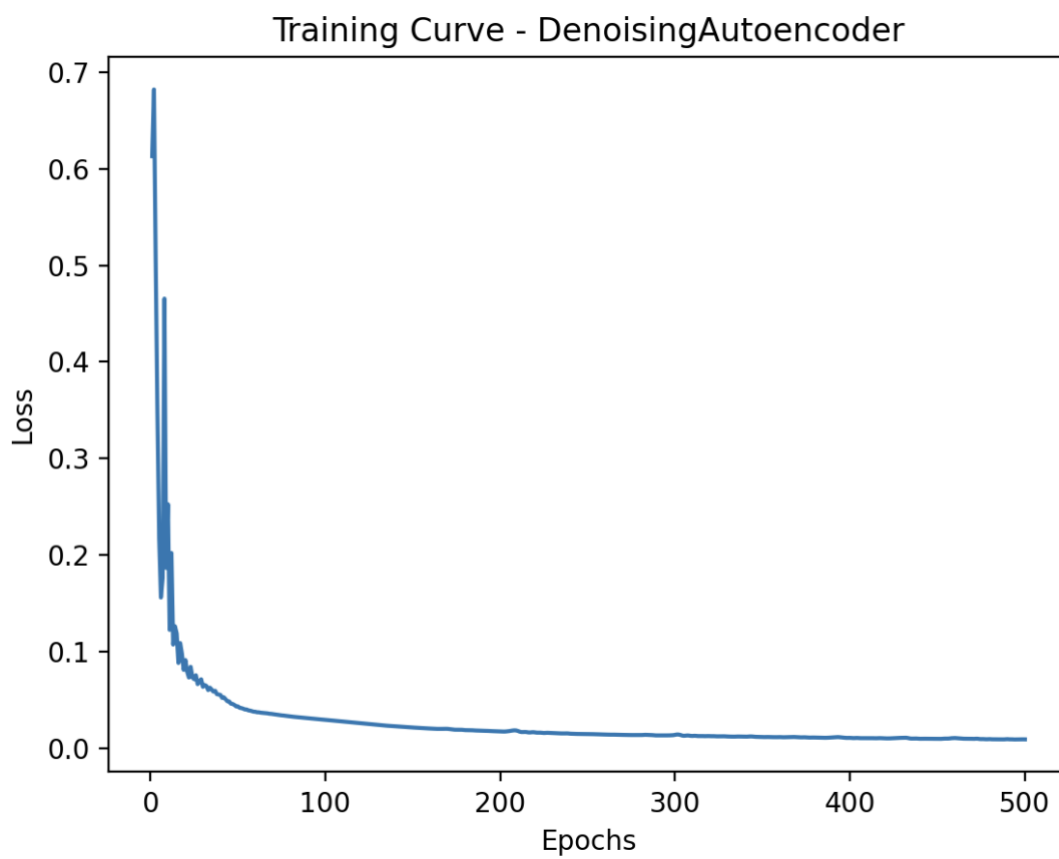
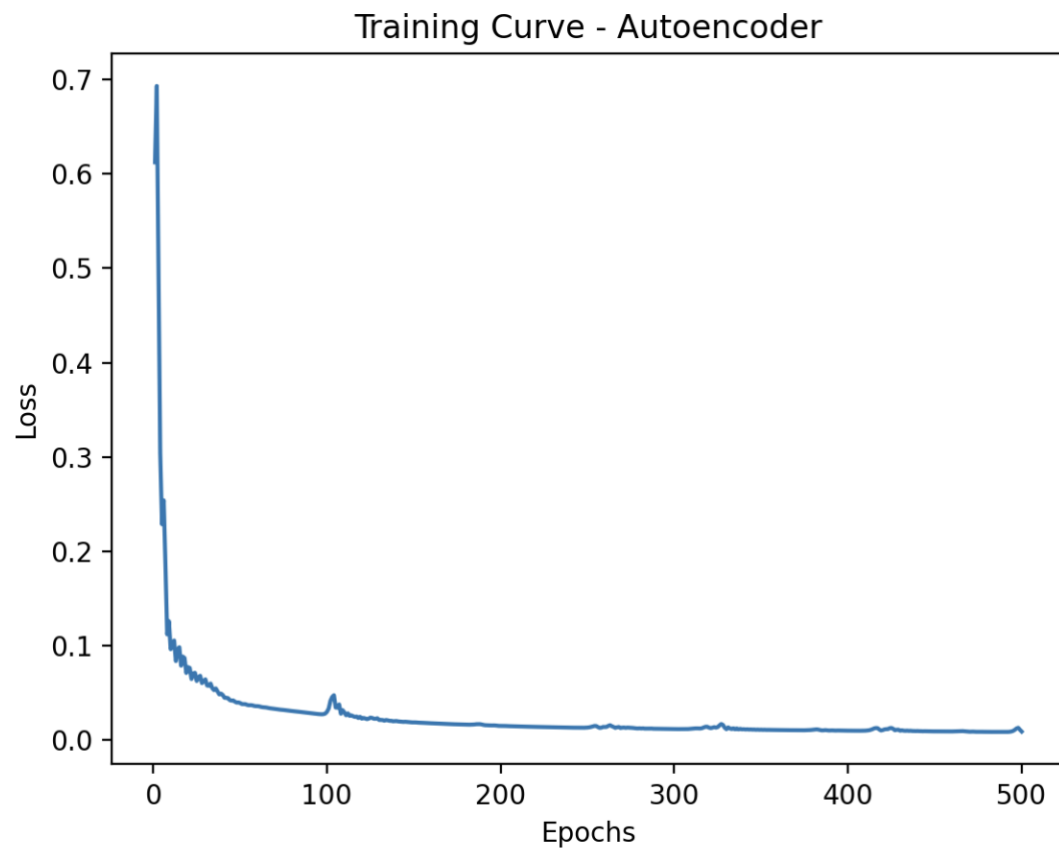


(b)

MSE: PCA = 0.0107, Autoencoder = 0.0148, Denoising Autoencoder = 0.0138



(c)



(d)

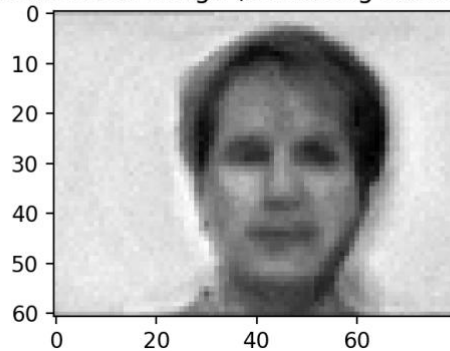
Shallower:

Encoder: Linear(4880, 488) -> Linear(488, 244) -> Linear(244, 122)

Decoder: Linear(122, 244) -> Linear(244, 488) -> Linear(488, 4880)

MSE: 0.0137

Reconstructed Image (Denoising Autoencoder)



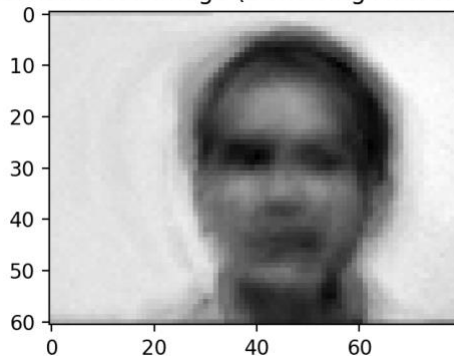
Deeper:

Encoder: Linear(4880, 1952) -> Linear(1952, 488) -> Linear(488, 244) -> Linear(244, 122)

Decoder: Linear(122, 244) -> Linear(244, 488) -> Linear(488, 1952) -> Linear(1952, 4880)

MSE: 0.0191

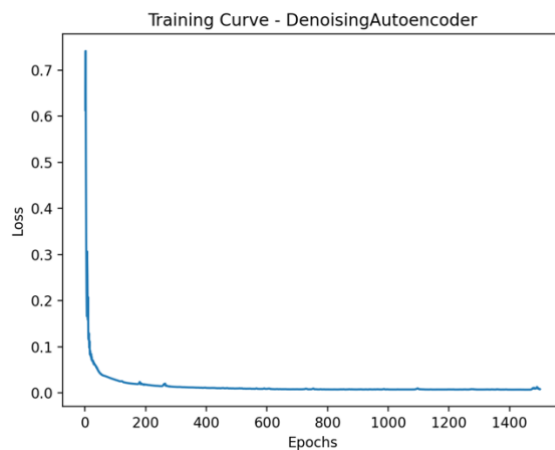
Reconstructed Image (Denoising Autoencoder)



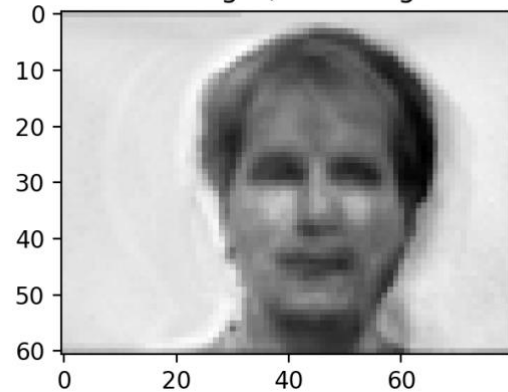
We can see that deeper model does not perform better in this case. The reconstruction error for shallower model is lower.

(e)

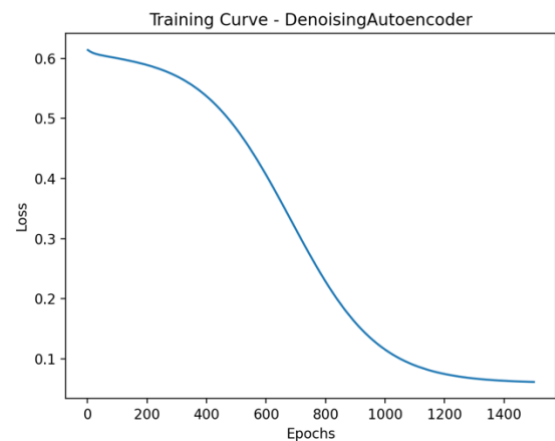
Adam:



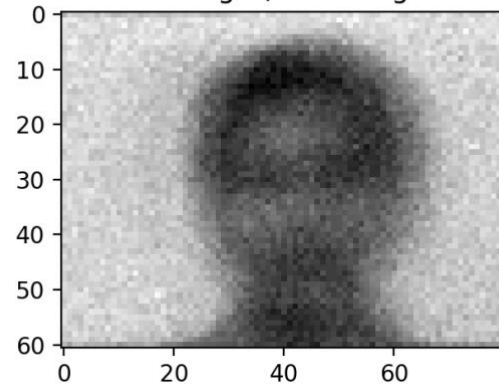
Reconstructed Image (Denoising Autoencoder)



SGD:



Reconstructed Image (Denoising Autoencoder)



From the training curves above, we can see that Adam has faster convergence speed and is able to find better local minimum; the reconstructed image also tells us that Adam(mse = 0.0132) performs better than SGD(mse = 0.0374).